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# A Causal Model to predict the Effect of Business Process Evolution on Quality of Service

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## ABSTRACT

Managing *Quality of Service* (QoS) of Service-based systems is a key challenge to produce systems that fulfill their requirements. Verifying the respect of a QoS contract in a system becomes more and more difficult as systems are more and more complex. Moreover, systems have to evolve in order to fulfil constantly changing requirements. As QoS properties are influenced by hidden factors such as connection rate or the system execution itself, determining the cause of a performance degradation is not mainstream. We propose in this paper to identify the causal relations to make explicit the hidden factors of influence. We more specifically focus on the consequences of system evolution with respect to QoS properties: using causal relations, we aim at predicting the possible overhead caused by an evolution. This paper shows through an example of Business Process how our evolution analysis helps to understand the effect of evolution on QoS property such as the Response Time. We show its efficiency by comparing the prediction with measured values.

## Categories and Subject Descriptors

D.2.5 [Software Engineering]: Testing and Debugging—Monitors; D.2.9 [Software Engineering]: Management—Life cycle; D.2.4 [Software Engineering]: Software/Program Verification

## Keywords

Business Process; Evolution; Quality of Service; Causal Model;

## 1. INTRODUCTION

*Quality of Service* (QoS) is one of the main concerns in building and processing software. We consider in this paper QoS as “the measurable, quantifiable properties of a software product” [7]. An example of such property is the *Response Time* (RT), defined as “the time spent between the reception of a request and the sending of the reply”. A QoS property characterises the system, or a part of the system. It is possible

to determine the QoS of software by static analysis at design time, or by monitoring the system at runtime [2]. This analytical process produces a measure, called *Property Value* (PV). However, even if the QoS property value is known, understanding the meaning of a *PV* is complex due to the presence of hidden factors of influence. In the case of Response Time, the QoS property value of a given activity fluctuates considerably depending on system elements (e.g., input of a service, coding style) or on external factors such as the number of simultaneous connections, the congestion of the network or the number of exchanged messages [14]. The causal relations among elements of the system, or with external factors are not explicit, leading to a misunderstanding of the system behaviour.

*Service Oriented Architectures* (SOA) define maintainable systems around the concept of *service*. Services can be composed in a *Business Process* (BP), an ordered set of activities, to implement larger services [6]. We consider *evolution* of a SOA system as the process of changing the state of a system to fulfil new user requirements. Evolutions can be local (e.g., by only changing a parameter of a service call). However, independently of how small the modification is, an evolution has consequences on the QoS of the entire system. Moreover, it is difficult for the BP architect to predict the QoS of a Business Process composed by hundreds of activities.

The contribution of this paper is to explicit the hidden causal relations between BP elements *w.r.t* QoS properties. We define *causal relations* [21], helping the BP architect to control the effect of an evolution on the QoS of a Business Process. Causal relations are part of our tool named SMILE (*Service Modeling for Impact of evolution framework*), that analyses QoS properties at design and runtime.

The remainder of the paper is as follows: Section 2 motivates our proposal with the presentation of our running example and the resulting challenges. Section 3 introduces causal modeling. In Section 4, we present a first causal model based on the execution of the system, and show its usage to perform an evolution analysis. Section 5 details a QoS-enrichment of the causal model to improve the evolution analysis. We validate the approach in Section 6 and compare it with related work on Section 7, before concluding with a summary and future work.

## 2. MOTIVATION AND CHALLENGES

### 2.1 Running Example

PICWEB is a Business Process that follows SOA methodological guidelines [20], positioning it as a typical use of SOA

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Figure 1: Evolution of PicWEB

for experimental purposes. This BP is a sub-part of a larger legacy system that is deployed and daily used in three institutions (70kLoC). Its implementation relies on partners such as FLICKR (Yahoo!) or PICASA (Google) to store and retrieve pictures. It allows one to retrieve a set of pictures with a given keyword from existing partner services.

FIG. 1.(a) depicts the initial version of PICWEB using an UML activity diagram. At the beginning, the BP fetches pictures matching a given keyword. The BP receives the keyword, invokes the PICASA service to retrieve matches, format them to have pictures with an appropriate dimension to fit the display screen, and returns them as result. During its life-cycle which started eight years ago, PICWEB evolved several times to respond to changes in users requirements.

In the scope of this paper, we focus on the following evolution, depicted in FIG. 1. PICWEB evolved to support an alternative picture provider (the first one was PICASA), as shown in FIG. 1(b). Such provider is called in parallel of the PICASA one. Their results are concatenated in the reply. After making the evolution, not only PICWEB but the entire system slowed down, forcing to stop and re-deploy an older version of the system. If the reason of the performance degradation was obviously the evolution, it took time for the BP architect to understand why and how the modification affected the entire system.

## 2.2 Challenges

The goal of this paper is to present an approach insuring a QoS-safe evolution of Business Processes. In other words, the BP architect must control that an evolution will not have unplanned effects on the overall system performance. To properly maintain QoS during the evolution process and to avoid QoS degradation, the BP architect needs to be supported. This consists into the following challenges:

1. How to ensure the preservation of a given QoS property while making a BP evolve? (See Section 3)

2. How to identify the system elements that have been affected by an evolution? (See Section 4)
3. How to model a QoS property in order to identify relevant hidden causal relations? (See Section 5)

To reach this goal, we propose an analysis that makes explicit the hidden causal relations inside a Business Process. The analysis helps to predict the chain reaction of an evolution on the QoS property values.

## 2.3 Assumptions

We assume in this paper that a *QoS expert* (in charge of the study of QoS properties) cooperates with a *BP architect* (responsible of the design of the BPs). We focus here on QoS properties that are composable, quantifiable, and defined on a single dimension. Where models like Palladio [3] focus on performance analysis, we assume here that the BP architect studies the QoS of the system without assessing any performance model.

## 3. A CAUSAL MODEL TO ANALYSE THE EFFECT OF AN EVOLUTION

In this section, we motivate the need of a causal model. We first introduce the concept of evolution and explain why a causal analysis is needed. Then, we present the model of our system that is finally analysed to build the causal model.

### 3.1 Definition of an Evolution

Several definitions in the literature introduce the term of *evolution*. The first to introduce it was Lehman [13]. Calinescu *et al.* consider evolution as “*changes in the environment or the specification causing changes in the implementation*” [5]. In the context of this paper, we call *BP evolution* an ordered set of changes made on the BP such as the add, deletion or update of activities (and/or their inner elements). But analysing the consequences of an evolution is not as mainstream as expected. In fact, checking the QoS of activities manipulated during the evolution is not sufficient. The common way to deal with the evolution is to re-check the entire system for each QoS property. With BPs made of hundreds of activities, it takes a considerable time to do such a re-check.

We propose an alternative method to determine the effect of an evolution on QoS. This method relies on the computation of a *Causal Model* representing the causal effects of a change in a system state. The evolution analysis uses the causal model to determine which elements of the system have been affected by the evolution. With this approach, we reduce the number of re-checks to perform on the system by considering only causally related elements. A full re-check of the system is no longer necessary. We will see in the following what a causal model is and how to define it.

### 3.2 Definition of the Universe

Performing an evolution consists in the modification of elements of the *Universe*. We consider here that a *Universe* represents the entire ecosystem of the SOA application. It is composed of three kinds of element:

- *System Elements* - It includes every element composing the system, *e.g.*, BPs, activities, variables. Activities represents the reception of a message, the invocation of a service, or any task involved in the definition of a

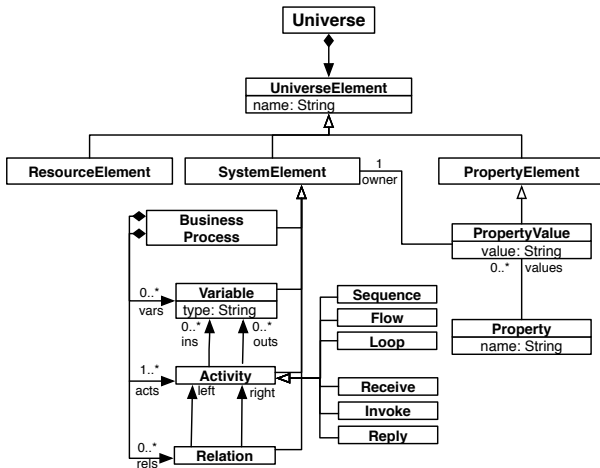


Figure 2: Extract of the Universe Meta-Model

BP. Variables are data defined locally in the BP that are used as input/output parameters in the invocation of a service.

- *Property Elements* - Every Property Value computed using the *QoS* property definition. A property value could be the measure or the application of a formula on a specific System Element (see Section 5.1).
- *Resource Elements* - The system uses resources (*e.g.*, memory, network). In this category, we represent every factor related to the system, such as the workload or the users.

FIG. 2 is an excerpt of our Universe meta-model. The System Elements are established by importing BPs into our tool. Property Elements store the value of a System Element for a given *QoS* property. They are positioned automatically by SMILE.

Each element of the *Universe* can act on other elements, causing a change in their states. It is necessary to model explicitly the actions of Universe elements on others in order to capture the entire effect of an evolution. Our evolution analysis relies on the Universe. By establishing the Universe elements directly involved in the evolution, and by following the causal relations starting from them, we isolate the Universe elements that have been impacted. Thus re-checking only the *QoS* of these elements is sufficient to determine the effect of an evolution, without re-checking the entire system.

### 3.3 Causal Model

To control the effect of an evolution on the system, it is necessary to make explicit the influence a Universe element can have on other elements of the Universe. For example, changing the content of the input message of a service influences its execution. We call such relation between elements *Causal Relations*. Let  $(A, B) \in \text{Universe Element}$ . We define a causal relation  $A \xrightarrow{x} B$ , the fact that if the state of A changes, then the state of B changes as well. Based on causal relations, we build a causal model that makes explicit causal relations of interest between elements of the Universe. A *Causal Model* is a digraph  $V(E, R)$  where E is the set of the Universe elements and R represents the set of causal relations applied to the BP. The evolution analysis uses the causal model to determine

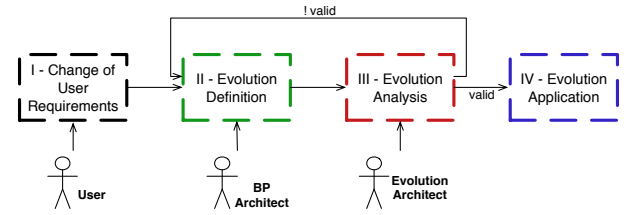


Figure 3: Overall Process to Evolve a System

the elements directly affected by a given evolution. Then, by following every causal relation starting from them (*i.e.*, by processing their transitive closure), the evolution analysis determines which elements have been indirectly affected.

Using this method, the Evolution architect can determine the set of affected elements, in order to re-check only their *QoS* property value. Our method restricts the number of elements to re-check and gives insights about the reasons of a possible performance degradation thanks to the chain of consequences represented by causal relations.

The following focuses on the causal model reasoning part of the evolution analysis (depicted in FIG. 6) to determine the set of BP elements whom had their *QoS* property values affected by an evolution. We detail the different steps of the method and illustrate them with the evolution of PicWEB.

## 4. A PROCESS TO ANALYSE THE EFFECT OF AN EVOLUTION

In this section, we present our evolution analysis that predicts the effect of an evolution on the *QoS* of a Business Process. The evolution analysis is an automated process that relies on a *Causal Model*. We introduce the overall approach of the evolution analysis, before introducing the causal relations extractable from the system execution, and applying the evolution analysis on PicWEB.

### 4.1 Overall Approach of Evolution Analysis

FIG. 3 depicts the overall process used to perform an evolution. The process is initiated by the *User*, who changes the requirements of the system. The *BP architect* defines a modification to make the system evolve. The proposed evolution is analysed: using the causal model, evolution analysis collects affected system elements and performs a *QoS* recheck of these elements to get the property value of the evolved Business Process. If the *Evolution architect* is satisfied, the evolution is accepted and the new version is deployed. Otherwise, the evolution is discarded and results of the analysis are used to design a better evolution.

#### 4.1.1 Evolution Definition

The SMILE Evolution Engine provides a set of actions to express BP evolution. This set of actions (which has been established empirically) is extensible, making possible to connect any evolution engine (such as ADORE [17] or PRAXIS [4]) with SMILE, as long as this evolution engine translates its behaviour into SMILE evolution actions. With our engine, it is possible to add or remove activity, variable, invocation parameters, or partial order. The following describes the effect of each evolution operation:

- *addActivity(act)/delActivity(act)*: this action adds/-deletes the activity *act* to the process.

```

1  addVariable(PicsPicasa)
2  delOutParameter(Picasa,Pics,0)
3  addOutParameter(Picasa,PicsPicasa,0)
4
5  addVariable(PicsFlickr)
6  addActivity(Flickr)
7  addInParameter(Flickr,Keyword,0)
8  addOutParameter(Flickr,PicsFlickr,0)
9
10 addActivity(Helper)
11 addInParameter(Helper,PicsPicasa,0)
12 addInParameter(Helper,PicsFlickr,1)
13 addOutParameter(Helper,Pics,0)
14
15 addRelation(receive,Flickr)
16 addRelation(Flickr,Helper)
17 addRelation(Helper,Format)
18 addRelation(Picasa,Helper)
19 delRelation(Picasa,format)

```

Figure 4: Script of the PICWEB evolution

- *addVar(var) / delVar(var)*: this action adds/deletes the variable *var* to the process.
- *addInPar(var,act,pos) / delInPar(var,act,pos)*: this action adds/deletes the variable *var* as an input parameter of the activity *act* at position *pos*.
- *addOutPar(var,act,pos) / delOutPar(var,act,pos)*: this action adds/deletes the variable *var* as an output parameter of the activity *act* at position *pos*.
- *addRelation(act1,act2) / delRelation(act1,act2)*: this action adds/deletes a partial order between activities *act1* and *act2*.

FIG. 4 is the script describing the evolution performed in FIG. 1. It is composed of four parts: lines 1-3 update the PICASA activity by changing its output by a new variable, *PicsPicasa*. Lines 5-8 add the FLICKR activity. Lines 10-13 are responsible of the call of the Helper activity, which concatenate *PicsPicasa* and *PicsFlickr* into *Pics*. Finally, lines 15-19 set up a new partial order, as depicted in FIG. 1.

#### 4.1.2 Evolution Analysis

The evolution analysis takes as input an evolution, and uses the causal model to compute the set of affected system elements. The causal model is the key of the analysis: it is initially defined by the application of the causal rules on the initial system, and enriched for each evolution. For each operation of the evolution, the analyser determines the system elements that are directly manipulated, *i.e.*, whose state is changed. Then, system elements that are affected by this change are collected by following causal relations present in the causal model. Performing the analysis of an evolution is an automated process composed of several tasks (described in Sect. 4.3):

1. MODIFICATION OF THE SYSTEM: The evolution is applied on the system to modify its structure. For each system element, SMILE insures the Universe consistency by creating every missing model element, like property

values. For example, adding the FLICKR invocation implies for SMILE to create a Response Time property value for the invoke activity. Moreover, because the determination of affected elements is a computation made on the causal model, SMILE has to update it.

2. DIRECTLY AFFECTED ELEMENTS COLLECTION: The evolution analysis takes as input the elements that are directly manipulated during the evolution. This step analyses the evolution operations to extract every system element manipulated by the evolution.
3. EVOLUTION CAUSAL ANALYSIS: Starting from the previously collected elements, the analysis browses the causal model to collect every *Indirectly Affected Element*. The analyser computes the transitive closure of causal relations starting from the Directly Affected Elements and collects every system element of the closure.
4. QoS RE-CHECK OF AFFECTED ELEMENTS: The causal analysis points out the affected elements. It is necessary to re-measure or to re-check them in order to get their new property values and to quantify the effect of the evolution. SMILE uses analysers described in the *QoS Property Definition* (see Sect. 5) and deploys a new version of the system to monitor the new property values.
5. EVOLUTION ARCHITECT DECISION: Based on the analysis results, the Evolution architect decides if the evolution has the desired effect and if he/she wants to effectively apply the evolution: (a) If the evolution does not produce a performance degradation, the system is deployed and can be used. (b) In case of performance degradation, the Evolution architect can choose to discard the evolution. The Business Process is restored to its state before the evolution, as well as the causal model. This is the only step of the process that requires the intervention of the Evolution architect.

## 4.2 Building the Causal Model of a System

We present in this part how the causal relations are determined for a given system. One way to define causal relations is to study the system, and to manually establish each causal relation that exists in the system. This method is time-consuming and error-prone. Instead, we introduce the notion of *Causal Rule*. A *Causal Rule* describes how to select under a specific condition the couples  $(a, b) \in UniverseElement$  that are causally related. For example, one kind of causal relation is the influence of an input on the processing of a service. The following rule describes how to build every relations of this kind for a given system:

$(VAR, ACT) \in (Variable \times Activity): VAR \in input(ACT)$   
*implies*  $VAR \xrightarrow{r} ACT$ .

Using this rule, SMILE analyses the Business Process to collect and build the different causal relations. We call these relations *Execution Engine Causal Relations*. This kind of rules is specific to the execution engine. We assume that they are provided with its specification, or written by an expert of the execution engine. In the same way, another rule collects variables and activities to represent the fact that an output is influenced by the execution of its activity. By applying these rules on PICWEB, SMILE generates the causal model. FIG. 5 depicts the causal model of PICWEB after evolution, obtained by application of the rules.



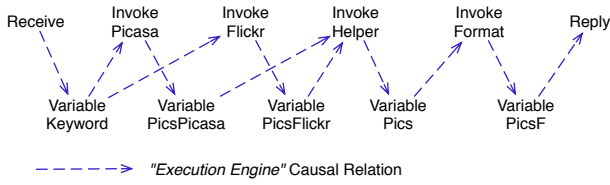


Figure 5: Causal Model of PicWEB after Evolution

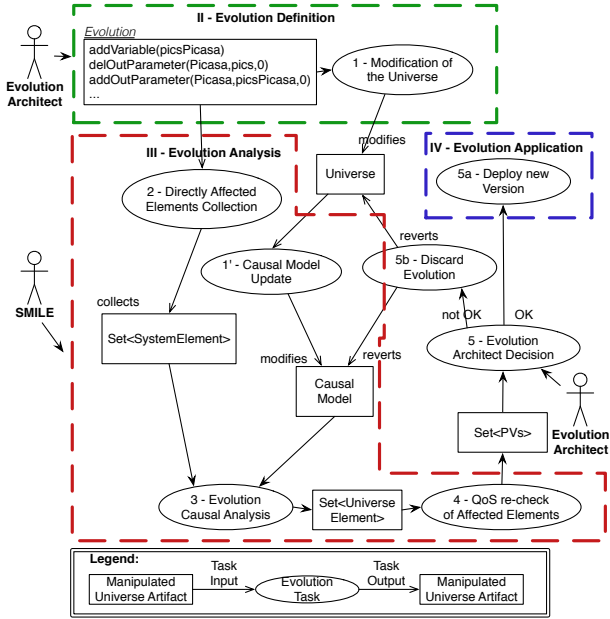


Figure 6: Overall Approach to Analyse an Evolution

The causal model defines the causal chains of the system. It is now possible, from one element, to obtain the causal chain of the changing of its state, and to collect which elements are indirectly affected by the change of state.

### 4.3 Application of the Evolution Analysis

We focus in this part on the evolution analysis, noted as step III in FIG. 3. The analysis reads the evolution script to collect system elements that are directly manipulated. From these elements, the evolution analysis uses the causal model to collect everything that is causally related and needs to be re-checked. It can be other system elements, but also property values or resources. FIG. 6 gives a closer look of steps II-IV of the evolution process. We explain in the following the different steps of the analysis, and apply it on PicWEB:

**Collection of Directly Affected Elements:**  
**Evolution**  $\rightarrow$  **Set<SystemElement>** -

SMILE analyses the evolution in order to collect the Directly Affected Elements. These elements are the starting points of the causal analysis, as these entities are the ones that have their state explicitly changed. Evolution operations can be classified into adding/removing operations. In the case of an add, a new system element is inserted, potentially having a causal effect on the system. The causal model must be checked from this system element. In the case of a deletion, a system element is removed. The effect this element had on the system is no longer established: as a consequence, if the element was in part responsible of other elements states,

Operation	Description	Collected Element
addActivity(act)	act is added to the <i>Universe</i>	act
delActivity(act)	act is removed from the <i>Universe</i>	act
addVar(var)	var is added to the <i>Universe</i>	var
delVar(var)	var is removed from the <i>Universe</i>	var
addInPar(act,var,pos)	var is added as a parameter of act	var
delInPar(act,var)	var is removed from act parameters	act
addOutPar(act,var,pos)	var is added in the return value of act	var
delOutPar(act,var)	var is removed from the return value of act	elements related to var

Table 1: Matching Between Universe and Causal Model Evolution Operations

removing the element cancels the action it had, thus causing a change in the state of previously affected elements. As a result, every previously causally related element must be re-checked as well. TAB. 1 regroups the rules used to collect DAEs. These rules have to be provided once by the Evolution architect if the Evolution Engine is extended.

In our example (FIG. 4), the evolution is composed of ten adding operations (lines 1, 3-13) and one deletion operation (line 2). Operations related to the order (lines 15-19) are not considered as we are not focused on the order of the execution. Considering the rules in TAB. 1, adding operations are analysed to collect *PicsPicasa*, *PicsFlickr*, *Flickr*, *Helper*, *Keyword*. In the case of the *delOutParameter* operation (line 2), analysis leads to the adding of every activity using variable *Pics*.

The unification of the analysis of the evolution operations produces the set  $\langle \text{PicsPicasa}, \text{PicsFlickr}, \text{Keyword}, \text{Flickr}, \text{Helper} \rangle$ . The analyser can determine from them which elements are potentially affected.

#### Evolution Causal Analysis

**Set<SystemElement>**  $\rightarrow$  **Set<UniverseElement>** -  
This step collects elements indirectly affected by the evolution. This task is done by following causal relations in the causal model, starting from each element of the Directly Affected Elements set. An example of causal relation modelled in the causal model is the influence of an input on the execution of a service (called serviceCall relations). The goal of this step is to use causal relations to identify system elements that are not manipulated by the evolution itself, but are affected due to a causal effect. By performing a transitive closure of the causal relations starting from directly affected elements, the analysis collects directly related elements, but also indirectly related elements.

From the variables of the DAE set, following causal relations allows the analyser to collect every activity for which a change in the input impacts the Response Time. As a result, the analysis collects several activities that are already in the DAE set (*Flickr*, *Picasa*, *Helper*), but also the *Format* activity: indeed, the *Helper* activity processes now a bigger number of pictures, which implies the Response Time of the *Format* activity to be different. Moreover, as Response Time of basic activities has changed, their aggregation is also affected. By following aggregation relations, the indirectly affected aggregation property values are *Flow*, *Sequence* and *PicWEB*.

### QoS PV re-check of Affected Elements

$\text{Set}\langle\text{UniverseElement}\rangle \rightarrow \text{Set}\langle\text{Property Value}\rangle$  - This step determines the new QoS property values of affected elements. SMILE equips the Business Process with targeted monitoring facilities and deploys it to capture the information. In the example, SMILE needs to monitor the Response Time of activities of the Indirectly Affected Element set. The monitored values show that globally, activities directly affected by the evolution does not take more time. However, the Format activity, which was not in the set of Directly Affected Elements, had a serious increase in its Response Time. We will see in the next section that this is due to the time-consuming Computation Time necessary to process each picture. The increase of the Computation Time caused the slowing down of the entire Business Process.

In this section, we have detailed how the causal model can be used by SMILE to predict the effect of an evolution on a given property value. By computing the transitive closure of the system elements directly related to the evolution, the evolution analysis identifies a subset of the system that is affected and that need to be re-checked. This identification brings a solution to challenge 2: the BP architect knows from the analysis what is the effect of the evolution on the overall QoS of the system. Our solution is fully automated to get from an evolution and the causal model of the BP if the system's performances have been degraded or not.

## 5. QOS ENRICHMENT OF THE CAUSAL MODEL

We presented in the previous section a causal model built upon the causal effect of the execution engine due to the assumption that executing a BP modifies the state of system elements. This approach is efficient to capture the minimal effect of the execution. However, the level of details is not enough to understand the causal effects at the level of a QoS property. We need also to represent the causal relations relating directly property values, in order to express effect of a property value. The execution engine itself represents the main causal effect, but external factors such as network, users, *etc.*, need to be considered as well to capture effects such as workload. In this section, we improve the causal model with QoS-specific relations. The goal of this enhancement is to capture a more precise set of elements affected by introducing QoS causal relations that show the effect of property values on other elements. This reduces the number of re-check and enables a better understanding of the effect of an evolution. First, we present how a QoS expert defines a QoS property. Then, we explain how QoS-specific causal relations can be deduced from the QoS Description. Finally, we show how the enhancement is applied on PicWEB, and its benefits.

### 5.1 Definition of a QoS Property

We consider in this paper *QoS* as a quantifiable metric that can be measured, determined by analysis, or computed through a mathematical formula, based on other property values [1, 7]. The mathematical formula is applied on a system element, to compute a *Property Value*. In SOA, a common way to determine the *QoS* property value of a Business Process is to compute the property values of its basic elements (such as invoke activities), and to compose them using an aggregation formula [18]. Thus the *QoS* expert needs to define a way to compute basic elements property value, and a formula

```

1 Property RT{
2   Unit : ms;
3   Range : positive;
4   BasicComputation: CT + TT;
5   ApplicationPoint:
6   Sequence is Sum(children);
7   Flow is Max(children);
8   Loop is k × children;
9 }
10
11 Property CT{
12   Unit : ms;
13   Range : positive;
14   BasicComputation: monitor;
15   IsInfluencedBy: Input(a);
16 }
17
18 Property TT{
19   Unit : ms;
20   Range : positive;
21   BasicComputation: monitor;
22   IsInfluencedBy: Network;
23 }

```

Figure 7: Expression of Response Time

to aggregate them. To express a *QoS* property, we define a *Domain Specific Language* (DSL). In the following, we detail each element of the description on an example, namely the Response Time. This property is a derived property, defined as the sum of the time to transmit messages (Transmission Time) and the time to execute the service itself (Computation Time). FIG. 7 is the description of Response Time provided by the *QoS* expert, written using our DSL. All the information must be declared once, and are then automatically applied by SMILE on any system to compute property values. They are usual information to manage the QoS property of a system.

A *QoS* property is defined as:

DECLARATION OF THE PROPERTY UNIT AND RANGE

Since *QoS* property values are numeric values, they must be characterised by a unit and a range. Unit of Response Time is defined as an amount of *milliseconds* which is always positive.

DETERMINATION OF *QoS* PROPERTY VALUES FOR BASIC ELEMENTS

The *QoS* processing technique is composition-based, that is to say that property values are computed at the lowest level, and then composed for higher level Universe elements. For example, the size of a message is computed by composition of the size of its inner variables. In this case, variables are Basic Elements. There are three ways to compute a *PV* for Basic Elements: **i) static analysis at design time.** Static analysis analyses the BP at design time to compute *PVs*. This kind of technique is well-suited for properties like reachability or liveness [9]. **ii) monitoring at runtime.** In the case where information needed to compute *PVs* are not available at design time (*e.g.*, the number of elements of an array), monitors are positioned to observe the values at runtime. This kind of technique is well-suited for properties like Message Size or Availability, which need information only observable at runtime [23]. **iii) computation based on other property values.** Some *QoS* properties are called *Derived Properties* [7]. These properties are defined as a

	Response Time
Sequence	$\sum RT(Sequence.children)$
Flow	$\max(RT(Flow.children))$
Loop	$k \times RT(Loop.activity)$

Table 2: Response Time Aggregation Formulæ

mathematical function of other sub-properties. The property values are computed based on a composition formula. For example, the Response Time of an activity A can be defined as  $RT(A) = TT(A) + CT(A)$ . TT stands for *Transmission Time*, the time spent to send the request and to receive the reply. CT is the *Computation Time* of the service called.

The declaration of static analysers or monitors enables SMILE to automatically perform the analysis, or to position monitors in the system. In our case, RT is defined as a derived property, and CT and TT are measured using provided monitors.

#### DETERMINATION OF QoS PROPERTY VALUES FOR COMPOSITE ELEMENTS

To compute a *QoS PV* of a composite activity, the *QoS* expert describes an aggregation formula, that can be found in literature [18] (reminded in TAB. 2). Such formula computes a *PV* for a composite activity, based on the *PVs* of its inner elements. For example, the *QoS* expert defines in *ApplicationPoint* the aggregation formulae for Sequence, Flow and Loop composite activities. It is important to notice that not every *QoS* property fits the aggregation principle, as stated in [7].

#### INFLUENCE OF OTHER FACTORS

The previously defined formula gives a way of computing a *Property Value (PV)* for each element of the system. However, this formula does not make explicit that a *PV* can be influenced by external factors. Modelling that a specific resource influences a *PV* enables to include this in our evolution analysis. Here, the *QoS* expert writes other factors dependencies of the *PV*. For example, the *QoS* expert defines that Transmission Time of an activity is influenced by the delay of the network.

## 5.2 Extraction of QoS-Specific Causal Relations

A QoS property can be computed based on the property value of the inner elements, the architecture of the software, sub-properties, the usage profile and the system environment [7]. These elements influence the *QoS PV* and should be represented in the causal model. In fact, such causal relation is implicit in the causal model presented in section 4: collecting system elements and re-checking them means to re-compute all their *PVs*. In this part, we propose a finer-grained modelling approach to represent QoS-specific causal relations, in order to avoid a complete re-check of a system element's *PVs*. We present this improvement in two steps: first, we present the existing causal relations among property values. Then, we explain the relations there is between system elements and property values. Finally, we apply the presented causal relations on PicWEB.

TAB. 3 depicts the mapping between Property Kind [7] and our QoS Property Description, and illustrates it. For example, a *Derived Property* like Response Time depends on other properties (CT and TT). This information is provided in the QoS Property Description, in the *BasicComputation* section. SMILE uses it to generate causal relations such as  $TT(Act)$

Property Kind	QoS Property Description	Causal Relation
Directly Composable	ApplicationPoint	$RT(Invoke) \rightarrow RT(Sequence)$
Architecture-Related		Execution Engine Relations
Derived	BasicComputation	$TT(Act) \rightarrow RT(Act)$
Usage-Dependent	Not supported	
System Environment	InfluencedBy	$input(Act) \rightarrow CT(Act)$

Table 3: Collection of Factors of Influence deduced from the QoS Property Description

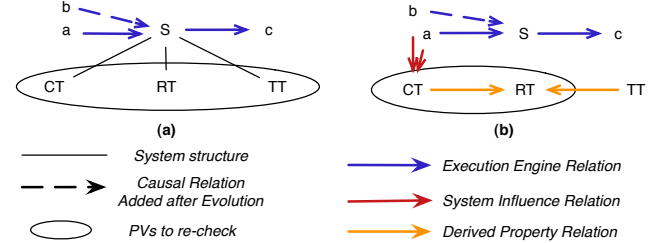


Figure 8: (a) Original Causal Model - (b) QoS-Enriched Causal Model

$\rightarrow RT(Act)$ , meaning that the RT of Act is influenced by its TT. SMILE supports four cases. We do not support *Usage-Dependent Properties*, because the Usage Profile has to be provided manually for each activity. With our approach, we wanted a *QoS* Description to be as independent of the system as possible. A Usage-Dependent Property would require a consequent additional modelling effort from the BP architect.

The mapping between QoS description and causal relations is done automatically by SMILE: the QoS expert describes the property, and SMILE uses it to generate the *PVs* and to get the QoS-specific causal relations of the system. The causal model is enriched with QoS-specific causal relations, that can be used to tune the evolution analysis and produced a more precise set of affected elements.

## 5.3 Enhanced Evolution Analysis

Using the previous causal relations, SMILE can build a new causal model that considers relations to *PVs*. With this enhanced model, we reduce the number of re-checks and improve the preciseness of the causal model. The causal model is now composed of relations between elements of the entire Universe. To illustrate the enhancement, we consider the defined RT on a simple call to a service *S*, taking a variable *a* as input, and producing a result stored in variable *c*. The Evolution architect decided to evolve this service call by adding another parameter as input, *b*. FIG. 8(a) is the causal model version 1, FIG. 8(b) is the QoS-enriched causal model. While evolution analysis of version 1 collects *S* and *c*, causing to re-check every QoS property value of *S*, the enhanced evolution analysis collects directly the CT and RT property values, avoiding to re-check the TT of *S*. This avoids to re-check unnecessary property values, and enables via the chain reaction to understand that the cause of a possible RT degradation would be the CT, and not the TT.

As a result, instead of re-checking every *PV* of an affected system element, the enhancement proposes to analyse only the *PVs* affected. In this way, the number of elements to re-check is reduced. Moreover, it enables a better understanding of the effect of an evolution, by pointing out directly which QoS property is affected. As an illustration, FIG. 9 depicts



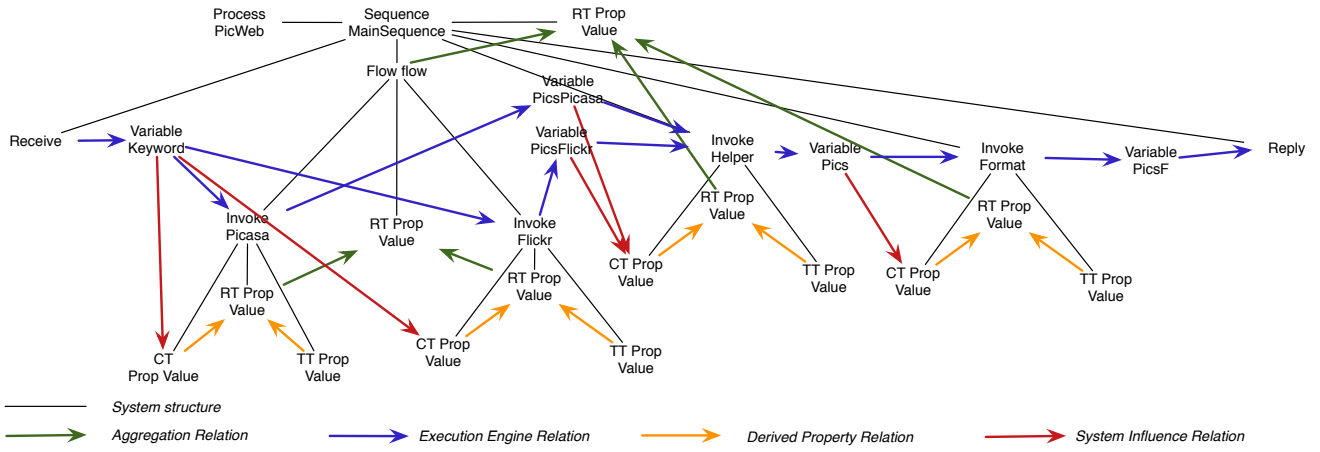


Figure 9: QoS-Enhanced Causal Model of PicWEB

the application of the causal rules extracted from the QoS Property Description to PicWEB.

In this section, we presented an improvement of the previous causal model to give more information about the causal relations among QoS property values and to reduce the number of re-checks. However, if this method gives more details about causality in a system, it also increases the size of the model and the execution time of the evolution analysis.

## 6. VALIDATION

In this section, we evaluate our approach empirically on a Crisis Management System (CMS) [11]. The goal of the experiment is to compare the different kinds of evolution analysis on a more consequent system.

### 6.1 Experiment

The Crisis Management System has been designed to help rescuers in the task of handling a crisis. The system is implemented by a set of four Business Processes and thirty-three services, positioning it as a representative example of a real-life system. Ten evolutions have been applied on the system to handle different extensions of the use cases. We took these evolutions as input of our evolution analysis. The goal of the experiment is to show the efficiency of our evolution analysis compared to traditional methods, based on a real case study. We performed the evolutions of the different Business Processes and analyse them.

The CMS is implemented in BPEL and executed on a SCA platform, namely FRASCATI<sup>1</sup> [22], using the Intent mechanisms to dynamically monitor its activities. Intents are codes that are weaved on a service call to be executed before, in parallel or after the service call. We developed a Response Time Intent that takes the current time when a request is received ( $t_1$ ), and after the reply is sent ( $t_2$ ). The intent deduces the Response Time of the invocation by subtracting  $t_1$  to  $t_2$ . Our experiment consists into the simultaneous handling of a set of three crisis, triggered by a call from three witnesses. The different version of the system are monitored to capture the response time of the activities.

### 6.2 Interpretation of the Results

We compared four methods: - the re-checking of the entire system, - the checking of elements directly affected by the evolution only, - the evolution analysis based only on system rules, and - the complete evolution analysis based on the causal relations extracted from the QoS Property Description.

TAB. 4 regroups the results of the experiment. We collected the number of property values that have to be re-checked in each case. There are several lessons to learn about this experiment: **Full Re-check is resource-consuming.** As stated in the table, the full re-check is the method having the most PVs to re-check. Moreover, the bigger the Business Process is compared to the evolution, the bigger the difference between this method and the others will be. On the other hand, this method insures not to forget anything. **QoS enhancement provides better results than system causal model.** As the QoS enhancement checks only the Computation Time and re-computes the Response Time, there is 1/3 less PVs to re-check for each activity. **Most of the time, the evolution analysis based on the system is as efficient as checking the evolution only.** This fact is due to the configuration of our experiment. Most of the evolution performed are "self-contained", in the sense that they do not affect variables that are re-used in the original Business Process. However, it is important to notice that the *UnavailableIntResource* is not self-contained: the evolution has a causal effect out of the scope of the evolution elements. As a result, the causal analysis produces results that are bigger than the re-check of the evolution, but they do include system elements that have been affected. **Checking only the evolution is not sufficient.** As stated in the previous point, the re-check of the elements of the evolution can forget causal effects, leading to a mistake in the prediction. In general, our methods provide better results than a full re-check, and similar or better results than a re-check of the evolution, with a better accuracy.

### 6.3 Threats to Validity

The experiment<sup>2</sup> showed that our prediction tool effectively determined activities affected by the evolution on our case study. We showed that the more the causal model is explained, the more accurate the analysis is. However, the accuracy has

<sup>1</sup><http://frascati.ow2.org>

<sup>2</sup>available at <http://smile.gforge.inria.fr/experiment.html>

BP / evolution	Full Recheck		Evol Recheck		System Causal Model		QoS Causal Model	
	Monit.	Comp.	Monit.	Comp.	Monit.	Comp.	Monit.	Comp.
<b>BP resolveACrisis</b>	7 acts							
MustAuthenticate	20	10	6	3	6	3	3	3
<b>BP handleACrisis</b>	5 acts							
ReHandleOnChange	20	10	10	5	10	5	5	5
<b>BP handleAMission</b>	11 acts							
UnavailableIntResource	24	12	2	4	<b>8</b>	<b>4</b>	<b>4</b>	<b>4</b>
UnavailableExtResource	30	15	8	4	8	4	4	4
ReHandleOnChange	32	16	10	5	10	5	5	5
<b>BP handleAWorker</b>	10 acts							
UseHelicopter	26	13	6	3	6	3	3	3
LostContact	26	13	6	3	6	3	3	3
Timeout	34	17	14	7	14	7	7	7
MissionFailed	26	13	6	3	6	3	3	3
ReHandleOnChange	30	15	10	5	10	5	5	5

Table 4: Number of elements to re-check for each analysis method

a cost: it is necessary for the QoS expert to give extra details about the definition of the QoS property. Moreover, the level of detail of the causal model leads to an increase in its memory size, which may have to be taken into consideration in the case of systems composed of hundreds of activities. Last but not least, the accuracy of the evolution analysis relies also on the correctness of the QoS Description. If some causal relations were not taken into consideration, it may be the case that the set of affected elements may differ from reality. We recommend to prefer an analysis encompassing too much elements than an analysis that could forget a causal effect that would lead to a non-QoS-safe evolution.

## 7. RELATED WORK

**Causal approaches** - Causal relations among elements of processes have already been studied. Luckham et al. introduced the notion of causality to study Software Architectures [15]. Mayerl et al. established metrics dependencies, with a specific focus on performance properties [16]. Their approach is very similar to ours. However, their work does not focus on the evolution of a BP. Other work focus on impact analysis of software evolution. For instance, Elbaum *et al.* present an empirical study on the effect of software evolution on code coverage testing [8]. They show that even a small modification of the software can impact code instructions involving to retest the code. We aim at applying the same approach on BPEL processes at runtime.

**Performance prediction** - Other QoS prediction techniques have been previously presented. The Palladio component model [3] is specialised into performance prediction. It defines a model of component behaviour and resource usage using parametric formula to estimate the QoS level of a system by simulation. If this approach is more precise in terms of property values due to their strong system and statistical models, we aimed with SMILE at reducing as much as possible the modelling effort without the task of writing several models such as performance, resource usage, *etc.* In our approach, only the QoS Description has to be written. Still, SMILE is precise enough to determines the *PVs* to re-check. Moreover, SMILE has been designed to

analyse evolution, which was not the intent of the Palladio Component Model. We could combine the preciseness of SMILE with Palladio, by importing Palladio’s resource QoS model, and generated a causal model from it. Similarly, the MARTE UML profile [19] can be used for performance modelling. It provides the Software architect with facilities to model time, resources and causality to perform performance analysis. However, as Palladio, it requires the definition of several performance models to predict efficient property values, and is not oriented on evolution analysis. Koziol et al. present the Q-Impress methodology which partially relies on Palladio and evaluate it on an industrial case [12]. They study code legacy to have a model of the system. Using it, they annotate components with their resources usage. These additional information help them to evaluate different architectural evolution strategies considering quality attributes. If their approach is more accurate in terms of QoS prediction than ours, we are more time efficient since we reduce the number of elements of the system to recheck while they recheck the entire system. Ivanovic et al. determine both at design time and runtime if Service Level Agreements (SLA) of an orchestration can be violated [10]. As us, they focus on QoS metrics that are measurable and can be aggregated. They express the different constraints as a Constraint Satisfaction Problem that they solve with constraint logic programming. This approach detects possible violations and determines the possible cause of such violation. However, it does not consider the evolution part of the development cycle.

## 8. CONCLUSION AND FUTURE WORK

We presented in this paper an approach to study the evolution of a Business Process with a specific focus on QoS. Our approach relies on the definition of the causal model of a BP, as a key to determine the effect of an evolution by identifying affected system elements. We defined an analysis of the evolution to predict its effect on the overall QoS property value, validated empirically by comparing the prediction with effective values and with other methods. Our approach allows a reduction of the re-checking of QoS property values

after evolution, and give clues about the origin of a QoS contract violation by depicting the chain of consequences of the evolution. We enable the possibility to qualify the effect of an evolution and to determine the cause of a potential overhead.

However, our approach is profitable as long as the time spent on computing the set of affected elements is lower than the time to verify the entire system. We need to determine the threshold under which our method creates an overhead compared to a complete re-check of the system. This challenge remains in our perspective. Moreover, we took the decision in this paper to study QoS property independently, without considering relationships between properties. Modelling these relationships would enable to build a hierarchy of properties and to detect if two properties are antagonist. Using this hierarchy of property would also help to propagate the effect of evolution among properties to avoid a systematic rechecking of every property. In the future, we will continue the development of SMILE to support QoS properties that are represented on several dimensions or that need a performance model to be studied. We will improve our tool to automate the property values determination and the selective monitoring. We also plan to use a probabilistic causal model to represent potential causality, using Bayesian Belief Network [21].

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